

Roll No.
28/09/21

Neeraj Appani TO 73

SDS

Page No.

Date

A1 Practical 10

Aim - Write a program to implement Gaussian Mixture Model

Algorithm:

1) Start

2) Import modules (matplotlib, scipy.stats, numpy)

3) use style five thirty eight

4) give random seed (8)

5) define x_0, x_1, x_2 to create creature data

6) Combine the cluster data

7) create array x with dimensionality $n \times 2$

8) Instantiate the random gaussians

9) Instantiate the random π_c

10) Probability for each datapoint x_i to belong to gaussian g

11) Normalize the probabilities such that each row of x sums to 1 and weight it by π_c
== fraction of points belonging to cluster c

12) to plot define columns in red, blue and green

13) plot the graph

14) Stop

AI 1o try.py - E:/fffiiles/college pracs and projects/AI/AI 1o try.py (3.8.3)
File Edit Format Run Options Window Help

```
import matplotlib.pyplot as plt
from matplotlib import style
style.use('fivethirtyeight')
import numpy as np
from scipy.stats import norm
np.random.seed(0)
#Neeraj Appari

X = np.linspace(-5,5,num=20)
X0 = X*np.random.rand(len(X))+10 # Create data cluster 1
X1 = X*np.random.rand(len(X))-10 # Create data cluster 2
X2 = X*np.random.rand(len(X)) # Create data cluster 3
X_tot = np.stack((X0,X1,X2)).flatten() # Combine the clusters

"""Create the array r with dimensionality nxK"""
r = np.zeros((len(X_tot),3))
print('Dimensionality','=',np.shape(r))

"""Instantiate the random gaussians"""

gauss_1 = norm(loc=-5,scale=5)
gauss_2 = norm(loc=8,scale=3)
gauss_3 = norm(loc=1.5,scale=1)

"""Instantiate the random pi_c"""
pi = np.array([1/3,1/3,1/3]) # We expect to have three clusters

"""
Probability for each datapoint x_i to belong to gaussian alpha
"""
```

Python 3.8.3 Shell

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Python 3.8.3 (tags/v3.8.3:6f8c832, May 13 2020, 22:20:19)
[MSC v.1925 32 bit (Intel)] on win32
Type "help", "copyright", "credits" or "license()" for more information.

>>>

===== RESTART: E:/fffiiles/college pracs and projects/AI/AI 1o try.py =====

Dimensionality = (60, 3)

[2.97644006e-02	9.70235407e-01	1.91912550e-07]
[3.85713024e-02	9.61426220e-01	2.47747304e-06]
[2.44002651e-02	9.75599713e-01	2.16252823e-08]
[1.86909096e-02	9.81309090e-01	8.07574590e-10]
[1.37640773e-02	9.86235923e-01	9.93606589e-12]
[1.58674083e-02	9.84132592e-01	8.42447356e-11]
[1.14191259e-02	9.88580874e-01	4.48947365e-13]
[1.34349421e-02	9.86565058e-01	6.78305927e-12]
[1.11995848e-02	9.88800415e-01	3.18533028e-13]
[8.57645259e-03	9.91423547e-01	1.74498648e-15]
[7.64696969e-03	9.92353030e-01	1.33051021e-16]
[7.10275112e-03	9.92897249e-01	2.22285146e-17]
[6.36154765e-03	9.93638452e-01	1.22221112e-18]
[4.82376290e-03	9.95176237e-01	1.55549544e-22]
[7.75866904e-03	9.92241331e-01	1.86665135e-16]
[7.52759691e-03	9.92472403e-01	9.17205413e-17]
[8.04550643e-03	9.91954494e-01	4.28205323e-16]
[3.51864573e-03	9.96481354e-01	9.60903037e-30]
[3.42631418e-03	9.96573686e-01	1.06921949e-30]
[3.14390460e-03	9.96856095e-01	3.91217273e-35]
[1.00000000e+00	2.67245688e-12	1.56443629e-57]
[1.00000000e+00	4.26082753e-11	9.73970426e-49]
[9.99999999e-01	1.40098281e-09	3.68939866e-38]
[1.00000000e+00	2.65579518e-10	4.05324196e-43]

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```

"""
Probability for each datapoint x_i to belong to gaussian g
"""
for c,g,p in zip(range(3),[gauss_1,gauss_2,gauss_3],pi):
    r[:,c] = p*g.pdf(X_tot) # Write the probability that x
                             # Therewith we get a 60x3 array f
"""
Normalize the probabilities such that each row of r sums to
cluster c
"""
for i in range(len(r)):
    r[i] = r[i]/(np.sum(pi)*np.sum(r,axis=1)[i])

"""In the last calculation we normalized the probabilities r
to belong to one gaussian (one column per gaussian). Since
to gaussian g, we have to do smth. like a simple calculation
x_i belongs to gaussian g. To realize this we must dive the
summing up each row in r and divide each value r_ic by sum(
look at the above plot and pick an arbitrary datapoint. Pic
belongs to this gaussian. This value will normally be small
the percentage that this point belongs to the chosen gaussi
gaussian divided by the sum of the probabilities for this da
point belong to each cluster c and threewith to each gaussi
assume that the points are equally distributed over the thr

print(r)
print(np.sum(r,axis=1)) # As we can see, as result each row
fig = plt.figure(figsize=(10,10))
ax0 = fig.add_subplot(111)

```

```

[9.99999878e-01 1.21709730e-07 1.17161878e-25]
[9.99999735e-01 2.65048706e-07 1.28402556e-23]
[9.99999955e-01 4.53370639e-08 2.60841891e-28]
[9.99999067e-01 9.33220139e-07 2.02379180e-20]
[9.99998448e-01 1.55216175e-06 3.63693167e-19]
[9.99997285e-01 2.71542629e-06 8.18923788e-18]
[9.99955648e-01 4.43516655e-05 1.59283752e-11]
[9.99987200e-01 1.28004505e-05 3.20565446e-14]
[9.64689131e-01 9.53405294e-03 2.57768163e-02]
[9.77001731e-01 7.96383733e-03 1.50344317e-02]
[9.96373670e-01 2.97775078e-03 6.48579562e-04]
[3.43634425e-01 2.15201653e-02 6.34845409e-01]
[9.75390877e-01 8.19866977e-03 1.64104537e-02]
[9.37822997e-01 1.19363656e-02 5.02406373e-02]
[4.27396946e-01 2.18816340e-02 5.50721420e-01]
[3.28570544e-01 2.14190231e-02 6.50010433e-01]
[3.62198108e-01 2.16303800e-02 6.16171512e-01]
[2.99837196e-01 2.11991858e-02 6.78963618e-01]
[2.21768797e-01 2.04809383e-02 7.57750265e-01]
[1.76497129e-01 2.01127714e-02 8.03390100e-01]
[8.23252013e-02 2.50758227e-02 8.92598976e-01]
[2.11943183e-01 2.03894641e-02 7.67667353e-01]
[1.50351209e-01 2.00499057e-02 8.29598885e-01]
[1.54779991e-01 2.00449518e-02 8.25175057e-01]
[7.92109803e-02 5.93118654e-02 8.61477154e-01]
[9.71905134e-02 2.18698473e-02 8.80939639e-01]
[7.60625670e-02 4.95831879e-02 8.74354245e-01]
[8.53513721e-02 2.40396004e-02 8.90609028e-01]]
[1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1.
1. 1. 1. 1. 1.
1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1.
1. 1. 1. 1. 1.

```

```

"""
Probability for each datapoint x_i to belong to gaussian g
"""
for c,g,p in zip(range(3),[gauss_1,gauss_2,gauss_3],pi):
    r[:,c] = p*g.pdf(X_tot) # Write the probability that x belongs to gaussian c in column c.
                             # Therewith we get a 60x3 array filled with the probability that each x_i belongs to one of
"""
Normalize the probabilities such that each row of r sums to 1 and weight it by pi_c == the fraction of points belongin
cluster c
"""
for i in range(len(r)):
    r[i] = r[i]/(np.sum(pi)*np.sum(r,axis=1)[i])

"""In the last calculation we normalized the probabilities r_ic. So each row i in r gives us the probability for x_i
to belong to one gaussian (one column per gaussian). Since we want to know the probability that x_i belongs
to gaussian g, we have to do smth. like a simple calculation of percentage where we want to know how likely it is in %
x_i belongs to gaussian g. To realize this we must dive the probability of each r_ic by the total probability r_i (thi
summing up each row in r and divide each value r_ic by sum(np.sum(r,axis=1)[r_i] )). To get this,
look at the above plot and pick an arbitrary datapoint. Pick one gaussian and imagine the probability that this datapo
belongs to this gaussian. This value will normally be small since the point is relatively far away right? So what is
the percentage that this point belongs to the chosen gaussian? --> Correct, the probability that this datapoint belong
gaussian divided by the sum of the probabilities for this datapoint and all three gaussians. Since we don't know how ma
point belong to each cluster c and threwith to each gaussian c, we have to make assumptions and in this case simply sa
assume that the points are equally distributed over the three clusters."""

print(r)
print(np.sum(r,axis=1)) # As we can see, as result each row sums up to one, just as we want it.

fig = plt.figure(figsize=(10,10))
ax0 = fig.add_subplot(111)

```



```

"""
Probability for each datapoint x_i to belong to one of the clusters
"""
for c,g,p in zip(range(3),[gauss_1,gauss_2,gauss_3]):
    r[:,c] = p*g.pdf(X_tot) # Write the probability for each point to belong to cluster c
    # Therewith we

"""
Normalize the probabilities such that each point belongs to one of the clusters
"""
for i in range(len(r)):
    r[i] = r[i]/(np.sum(r,axis=1))

"""In the last calculation we normalized the probabilities such that each point belongs to one of the clusters (one column per point). To realize this, we have to do smth. like a summing up each row in r and divide each row by the sum. To look at the above plot and pick an arbitrary point, we assume that the points are equally distributed. This value will be the percentage that this point belongs to this gaussian. This value will be the percentage that this point belongs to this gaussian divided by the sum of the probabilities of all points belonging to each cluster c and then we will assume that the points are equally distributed.

print(r)
print(np.sum(r,axis=1)) # As we can see,

fig = plt.figure(figsize=(10,10))
ax0 = fig.add_subplot(111)

```

